

Selecting a Minimax Estimator Doing Well at a Point*

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1. INTRODUCTION

Let $X_i \sim N(\theta_i, I_p)$, $i = 1, 2, \dots, n$ ($n > p + 1$, $p > 1$) be an independent finite sequence of p -dimensional vectors. Let $X = (X_1, \dots, X_n)$ be $p \times n$ matrix of random variables with the matrix of means $\theta = (\theta_1, \theta_2, \dots, \theta_n)$. Efron and Morris [2] generalized the James–Stein estimator: they proved that $X - (n - p - 1) S^{-1} X$ (along with its positive part $(I - (n - p - 1) S^{-1})_+ X$) is a minimax estimator. Stein [3] considered the estimators of form $X + \nabla \ln f(l)$, where $l = (l_1, l_2, \dots, l_p)^T$ is a vector with components $l_1 \geq l_2 \geq \dots \geq l_p \geq 0$, the characteristic roots of $S = XX^T$, and $\nabla \ln f(l) = ((\partial/\partial X_{jk}) \ln f(l))$ is the matrix of partial derivatives of the function $\ln f(l)$. He gave a condition under which the corresponding estimator $X + \nabla \ln f(l)$ is minimax. Zheng [4] gave a class of minimax estimators of the form $X + \nabla \ln f(l)$, where

$$f(l) = \prod_{i=1}^p \int_{l_i}^{+\infty} \frac{\mu_i(t_i) dt_i}{t_i^{1/2}(n-p+1)}. \quad (1.1)$$

But when $p > 1$, $n > p + 1$, the estimator $\delta(X) = X + \nabla \ln f(l)$, where $f(l)$ is given by (1.1), is inadmissible. For fixed $f(l)$ of form (1.1), Zheng [4] found a class of estimators of the form

$$\delta_\beta(X) = X + \nabla \ln f(l) - 2X \left/ \left(\frac{\text{tr } S}{p(p+1)-2} + \beta(l) \right) \right. \quad (1.2)$$

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where $\beta(l)$ is a function belonging to the set

$B = \beta(l)$:

- (i) $\sum_{i=1}^p (\partial/\partial l_i) \beta(l) \cdot l_i \leq 0$;
- (ii) $E_{\theta}\{(\partial/\partial l_i) \beta(l) \cdot \sqrt{l_i}\} < +\infty, i = 1, 2, \dots, p$;
- (iii) $\beta(l) \geq 0$, and $\beta(l) > 0$ on a set of positive measure.

For every $\beta \in B$, $\delta_{\beta}(X)$ improves upon the minimax estimator $\delta(X) = X + \nabla \ln f(l)$. Actually, we can slightly enlarge the class B without changing the original proof given in [4]. Let

$B_1 = \beta(l)$:

- (i) $E_{\theta}\{(\partial/\partial l_i) \beta(l) \cdot \sqrt{l_i}\} < \infty, i = 1, 2, \dots, p$;
- (ii) $(p(p+1)/2) \beta(l) - \sum_{i=1}^p l_i (\partial/\partial l_i) \beta(l) \geq 0$;
- (iii) $L \left\{ X: \frac{1}{2} p(p+1) \beta(l) - \sum_{i=1}^p l_i (\partial/\partial l_i) \beta(l) > 0 \right\} > 0$;
- (iv) $\beta(l) \geq 0$.

Then for $\beta(l) \in B_1$, the estimator $\delta_{\beta}(X)$ given by (1.2) improves upon the estimator $\delta(X) = X + \nabla \ln f(l)$.

The purpose of this paper is to select an estimator $\delta^*(X)$ which dominates $\delta(X) = X + \nabla \ln f(l)$ such that the estimator is the best at $\theta = 0$ among estimators $\delta_{\beta}(X) = \delta(X) - 2X/(\text{tr } S/(p(p+1)-2) + \beta(l))$, $\beta \in B_1$; i.e., δ^* satisfies

$$(i) \quad E_{\theta}\{\|\delta^* - \theta\|^2\} \leq E_{\theta}\{\|X + \nabla \ln f(l) - \theta\|^2\} \quad (1.5)$$

$$(ii) \quad E_{\theta=0}\{\|\delta^* - \theta\|^2\} = \inf_{\beta \in B_1} E_{\theta=0}\{\|\delta_{\beta} - \theta\|^2\} \quad (1.6)$$

where $\|\delta^* - \theta\|^2 = \text{tr}(\delta^* - \theta)^T (\delta^* - \theta)$, etc. Notice that δ^* does not have the form of δ_{β} for some $\beta \in B_1$. We will prove that δ^* has the form

$$\delta^*(X) = X + \nabla \ln f(l) - 2X \cdot g(X) \quad (1.7)$$

where

$$g(X) = \frac{1}{\text{tr } S} \cdot \min \left\{ \frac{1}{2} (\text{tr } S + \text{tr } \nabla \ln f(l)^T X)_+, p(p+1)-2 \right\}. \quad (1.8)$$

In (1.1), let $\mu_1(t) = \mu_2(t) = \cdots = \mu_p(t) = 1$. Then $\delta(X) = X + \nabla \ln f(l) = X - (n-p-1) S^{-1} X$, and

$$\delta^*(X) = X - (n-p-1) S^{-1} X - 2Xg_1(X) \quad (1.9)$$

where

$$g_1(X) = \frac{1}{\text{tr } S} \min \left\{ p(p+1) - 2, \frac{1}{2} (\text{tr } S - p(n-p-1))_+ \right\}. \quad (1.10)$$

In (1.1), let $\mu_1(t) = \mu_2(t) = \cdots = \mu_p(t) = \mu(t)$, where

$$\begin{aligned} \mu(t) &= 1 \quad \text{if } t > n-p-1 \\ &= (n-p-1)^{-(1/2)(n-p+1)} t^{(1/2)(n-p+1)} e^{(1/2)(n-p-1-t)} \\ &\quad \text{if } t \leq n-p-1. \end{aligned} \quad (1.11)$$

Then, $\delta(X) = (I - (n-p-1) S^{-1})_+ X$, and

$$\delta^*(X) = (I - (n-p-1) S^{-1})_+ X - 2X \cdot g_2(X) \quad (1.12)$$

where

$$g_2(X) = \frac{1}{\text{tr } S} \min \left\{ p(p+1) - 2, \frac{1}{2} \sum_{i=1}^p (l_i - (n-p-1))_+ \right\}. \quad (1.13)$$

2. THE MAIN RESULT

Suppose $\mu_1(t), \dots, \mu_p(t)$ are nondecreasing functions satisfying the condition of Theorem 1 of [4], i.e.,

(i) $\mu_{i+1}(t)/\mu_i(t)$ nondecreasing, $i = 0, 1, 2, \dots, p-1$,

$$\left(\mu \equiv 1, \frac{0}{0} \stackrel{\text{def}}{=} 1, \frac{1}{0} \stackrel{\text{def}}{=} +\infty \right). \quad (2.1)$$

$$(ii) \int_a^{+\infty} \mu_i(t) dt / t^{(1/2)(n-p+1)} < +\infty \quad (\forall a > 0), i = 1, 2, \dots, p. \quad (2.2)$$

Then according to Theorem 1 of [4], $X + \nabla \ln f(l)$ is a minimax estimator of θ , where $f(l)$ is given by (1.1). Now we have

THEOREM 1. Suppose that $\mu_1(t), \mu_2(t), \dots, \mu_p(t)$ are absolutely continuous, nondecreasing functions satisfying (2.1), (2.2) and that the functions of v ,

$$v - \frac{2\mu_i(v)v}{v^{(1/2)(n-p+1)}} \Big/ \int_v^{+\infty} \frac{\mu_i(t) dt}{t^{(1/2)(n-p+1)}} \quad i = 1, 2, \dots, p, \quad (2.3)$$

are nondecreasing functions of v . Then the estimator $\delta^*(X)$ given by (1.7) satisfies (1.5) and (1.6). In addition, we have the following:

$$\begin{aligned} E_{\theta=0} \{ \|\delta^*(X) - \theta\|^2 \} &= E_{\theta=0} \{ \|X + \nabla \ln f(l) - \theta\|^2 \} \\ &\quad - 4E_{\theta=0} \left\{ \frac{(\text{tr } S + \text{tr } \nabla \ln f(l)^T X) \text{tr } S^{-(1/2)(p+1)}}{\text{tr } S^{-(1/2)p(p+1)+1}/(p(p+1)-2) + \psi_0} \right. \\ &\quad \left. - \frac{\text{tr } S^{-p(p+1)+1}}{((\text{tr } S)^{-(1/2)p(p+1)+1}/(p(p+1)-2) + \psi_0)^2} \right\} \end{aligned} \quad (2.4)$$

where

$$\begin{aligned} \psi_0 &= +\infty, & \text{if } \text{tr } S + \text{tr } \nabla \ln f(l)^T X \leq 0; \\ &= \left(\frac{2 \text{tr } S^{-(1/2)p(p+1)+1}}{\text{tr } S + \text{tr } \nabla \ln f(l)^T X} - \frac{\text{tr } S^{-(1/2)p(p+1)+1}}{p(p+1)-2} \right)_+, \\ & & \text{if } \text{tr } S + \text{tr } \nabla \ln f(l)^T X > 0. \end{aligned} \quad (2.5)$$

Proof. Let

$$\Delta(\delta^*, \theta) = E_{\theta} \{ \|X + \nabla \ln f(l) - \theta\|^2 - \|\delta^* - \theta\|^2 \}. \quad (2.6)$$

First, we prove that (1.5) holds. Using (1.7), we get

$$\begin{aligned} \Delta(\delta^*, \theta) &= 4E_{\theta} \{ g(X) \text{tr } X^T(X - \theta) \} - 4E_{\theta} \{ g^2(X) \|X\|^2 \} \\ &\quad + 4E_{\theta} \{ g(X) \text{tr } X^T \nabla \ln f(l) \}. \end{aligned} \quad (2.7)$$

Using integration by parts,

$$E_{\theta} \{ g(X) \text{tr } X^T(X - \theta) \} = npE_{\theta} \{ g(X) \} + \sum_{j=1}^p \sum_{k=1}^n E_{\theta} \left\{ \frac{\partial g}{\partial X_{jk}} \cdot X_{jk} \right\}. \quad (2.8)$$

From (2.7) and (2.8), we get

$$\begin{aligned} \Delta(\delta^*, \theta) &= 4E_{\theta} \left\{ npg(X) + \sum_{j=1}^p \sum_{k=1}^n X_{jk} \frac{\partial g(X)}{\partial X_{jk}} \right. \\ &\quad \left. - g^2(X) \text{tr } S + g(X) \text{tr } X^T \nabla \ln f(l) \right\}. \end{aligned} \quad (2.9)$$

First, suppose that $0 < \frac{1}{2}(\text{tr } S + \text{tr}(\nabla \ln f(l)^\text{T} X)) < p(p+1) - 2$. Then, according to (1.8), $g(X) = (\text{tr } S + \text{tr} \nabla \ln f(l)^\text{T} X)/(2 \text{tr } S)$. Hence

$$\begin{aligned}
 & \sum_{j=1}^p \sum_{k=1}^n \frac{\partial g}{\partial X_{jk}} \cdot X_{jk} \\
 &= \sum_{j=1}^p \sum_{k=1}^n \frac{1}{2 \text{tr } S} \frac{\partial}{\partial X_{jk}} (\text{tr } S + \text{tr} \nabla \ln f(l)^\text{T} X) \cdot X_{jk} \\
 &\quad - \sum_{j=1}^p \sum_{k=1}^n \frac{\text{tr } S + \text{tr} \nabla \ln f(l)^\text{T} X}{2(\text{tr } S)^2} \cdot 2X_{jk}^2 \\
 &= \sum_{j=1}^p \sum_{k=1}^n \frac{1}{2 \text{tr } S} \frac{\partial}{\partial X_{jk}} (\text{tr } S + \text{tr} \nabla \ln f(l)^\text{T} X) X_{jk} \\
 &\quad - \frac{1}{\text{tr } S} (\text{tr } S + \text{tr} \nabla \ln f(l)^\text{T} X). \tag{2.10}
 \end{aligned}$$

From (4.24) of [3],

$$\text{tr} \nabla \ln f(l)^\text{T} X = -2 \sum_{i=1}^p \frac{\mu_i(l_i)}{l_i^{(1/2)(n-p-1)}} \Big/ \int_{l_i}^{+\infty} \frac{\mu_i(t) dt}{t^{(1/2)(n-p+1)}}, \tag{2.11}$$

so (using (4.24) of [3] again)

$$\begin{aligned}
 & \sum_{j=1}^p \sum_{k=1}^n \frac{1}{2 \text{tr } S} \frac{\partial}{\partial X_{jk}} (\text{tr } S + \text{tr} \nabla \ln f(l)^\text{T} X) X_{jk} \\
 &= \frac{1}{2 \text{tr } S} \sum_{i=1}^p \left(l_i - 2 \frac{\mu_i(l_i)}{l_i^{(1/2)(n-p-1)}} \Big/ \int_{l_i}^{+\infty} \frac{\mu_i(t) dt}{t^{(1/2)(n-p+1)}} \right)' \cdot 2l_i.
 \end{aligned}$$

Substituting the above into (2.10), we get

$$\begin{aligned}
 \sum_{j=1}^p \sum_{k=1}^n \frac{\partial g}{\partial X_{jk}} \cdot X_{jk} &= \frac{1}{\text{tr } S} \sum_{i=1}^p \left(l_i - \frac{2\mu_i(l_i)}{l_i^{(1/2)(n-p-1)}} \Big/ \int_{l_i}^{+\infty} \frac{\mu_i(t) dt}{t^{(1/2)(n-p+1)}} \right)' \cdot l_i \\
 &\quad - \frac{1}{\text{tr } S} (\text{tr } S + \text{tr} \nabla \ln f(l)^\text{T} X).
 \end{aligned}$$

So when $0 < \frac{1}{2}(\text{tr } S + \text{tr}(\nabla \ln f(l)^\text{T} X)) < p(p+1) - 2$, the following holds:

$$\begin{aligned}
 n p g(X) + \sum_{j=1}^p \sum_{k=1}^n X_{jk} \frac{\partial g}{\partial X_{jk}} - g^2(X) \cdot \text{tr } S + g(X) \text{tr } X^\text{T} \nabla \ln f(l) \\
 = g(X)(p(n-p-1) + \text{tr } X^\text{T} \nabla \ln f(l))
 \end{aligned}$$

$$\begin{aligned}
& + \frac{1}{\text{tr } S} \sum_{i=1}^p \left(l_i - \frac{2\mu_i(l_i)}{l_i^{(1/2)(n-p-1)}} \left/ \int_{l_i}^{+\infty} \frac{\mu_i(t) dt}{t^{(1/2)(n-p-1)}} \right. \right)' \cdot l_i \\
& + \frac{\text{tr } S + \text{tr } \nabla \ln f(l)^T X}{\text{tr } S} \left[\frac{p(p+1)}{2} - 1 - \frac{1}{4} (\text{tr } S + \text{tr } \nabla \ln f(l)^T X) \right] \\
& \geq 0.
\end{aligned} \tag{2.12}$$

Second, suppose that $\frac{1}{2}(\text{tr } S + \text{tr } \nabla \ln f(l)^T X) > p(p+1) - 2$. In this case, $g(X) = (1/\text{tr } S)(p(p+1) - 2)$. Then,

$$\sum_{j,k} \frac{\partial g}{\partial X_{jk}} \cdot X_{jk} = -2g(X).$$

So,

$$\begin{aligned}
& npg(X) + \sum_{j=1}^p \sum_{k=1}^n \frac{\partial g}{\partial X_{jk}} X_{jk} - g^2(X) \text{tr } S + g(X) \text{tr } X^T \nabla \ln f(l) \\
& = (p(n-p-1) + \text{tr } X^T \nabla \ln f(l)) g(X) \geq 0.
\end{aligned} \tag{2.13}$$

Finally, suppose that $\frac{1}{2}(\text{tr } S + \text{tr } \nabla \ln f(l)^T X) \leq 0$. In this case $g(X) \equiv 0$. Then,

$$npg(X) + \sum_{j,k} \frac{\partial g}{\partial X_{jk}} X_{jk} - g^2(X) \text{tr } S + g(X) \text{tr } X^T \nabla \ln f(l) \equiv 0. \tag{2.14}$$

From (2.12), (2.13), (2.14), and (2.9), we get (1.5).

To prove (1.6, we first prove the inequality

$$\begin{aligned}
\Delta(\delta_\beta, 0) & \leq 4E_{\theta=0} \left\{ \frac{(\text{tr } S + \text{tr } \nabla \ln f(l)^T X)(\text{tr } S)^{-(1/2)p(p+1)}}{((\text{tr } S)^{-(1/2)p(p+1)+1}/(p(p+1)-2) + \psi_0)} \right. \\
& \quad \left. - \frac{(\text{tr } S)^{-p(p+1)+1}}{((\text{tr } S)^{-(1/2)p(p+1)+1}/(p(p+1)-2) + \psi_0)^2} \right\} \quad \forall \beta \in B_1 \tag{2.15}
\end{aligned}$$

where ψ_0 is given by (2.5) and δ_β is given by (1.2).

According to Theorem 4 of [4], $\delta_\beta(X) (= X + \nabla \ln f(l) - 2X/(\text{tr } S/(p(p+1)-2) + \beta(l)))$ dominates the estimator $X + \nabla \ln f(l)$ and

$$\begin{aligned}
\Delta(\delta_\beta, \theta) & = 8E_\theta \left\{ \frac{(1/2)p(p+1) \beta(l) - \sum_{i=1}^p l_i (\partial/\partial l_i) \beta(l)}{(\text{tr } S/(p(p+1)-2) + \beta(l))^2} \right\} \\
& \quad + 4E_\theta \left\{ \frac{p(n-p-1) + \text{tr } X^T \nabla \ln f(l)}{\text{tr } S/(p(p+1)-2) + \beta(l)} \right\} \tag{2.16}
\end{aligned}$$

where $\beta \in B_1$. Considering the point $\theta=0$, since the distribution of $l=(l_1, \dots, l_p)$ is known [1], we get

$$\begin{aligned} \Delta(\delta_\beta, 0) = & \iint \cdots \int_{l_1 \geq \cdots \geq l_p \geq 0} \left[\frac{8((1/2)p(p+1)\beta(l) - \sum_{i=1}^p l_i(\partial\beta(l)/\partial l_i))}{(\text{tr } S/(p(p+1)-2) + \beta(l))^2} \right. \\ & \left. + \frac{4(p(n-p-1) + \text{tr } X^T \nabla \ln f(l))}{\text{tr } S/(p(p+1)-2) + \beta(l)} \right] \\ & \times \frac{\pi^{(1/2)p} \prod_{i=1}^p l_i^{(1/2)(n-p-1)} e^{-(1/2) \sum_{i=1}^p l_i} \prod_{i < j} (l_i - l_j)}{2^{(1/2)pn} \prod_{i=1}^p \{\Gamma(\frac{1}{2}(n+1-i)) \Gamma(\frac{1}{2}(p+1-i))\}} dl_1 \cdots dl_p. \end{aligned}$$

Let

$$\begin{aligned} u &= l_1 + \cdots + l_p, \\ \Delta_1 &= (l_1 - l_2)/u, \\ \Delta_2 &= (l_2 - l_3)/u, \\ &\vdots \\ \Delta_{p-2} &= (l_{p-2} - l_{p-1})/u, \\ \Delta_{p-1} &= l_{p-1}/u. \end{aligned} \tag{2.17}$$

Then,

$$\begin{aligned} \Delta(\delta_\beta, 0) = & \int \cdots \int \int_0^\infty \left[\frac{8((1/2)p(p+1)\beta(l) - \sum_{i=1}^p l_i(\partial\beta(l)/\partial l_i))}{(u/(p(p+1)-2) + \beta(l))^2} \right. \\ & \left. + \frac{4(p(n-p-1) + \text{tr } \nabla \ln f(l)^T X)}{u/(p(p+1)-2) + \beta(l)} \right] \\ & \times u^{(np/2)-1} e^{-u/2} du \cdot D(\Delta_1, \dots, \Delta_{p-1}) d\Delta_1, \dots, d\Delta_{p-1} \end{aligned}$$

where

$$\begin{aligned} \mathcal{D} = & \{(\Delta_1, \dots, \Delta_{p-1}): \Delta_i \geq 0, i=1, \dots, p-1, \\ & 0 < 1 - (\Delta_1 + 2\Delta_2 + \cdots + (p-1)\Delta_{p-1}) < \Delta_{p-1}\} \end{aligned}$$

and $D(\Delta_1, \dots, \Delta_{p-1})$ is a function of $\Delta_1, \dots, \Delta_{p-1}$. Denote $\beta(l) = r(u, \Delta_1, \dots, \Delta_{p-1})$, and it is easy to show that $\sum_{i=1}^p l_i(\partial\beta(l)/\partial l_i) = r'_u \cdot u$. So,

$$\begin{aligned} \Delta(\delta_\beta, 0) = & \int \cdots \int_{\mathcal{D}} D(\Delta_1, \dots, \Delta_{p-1}) d\Delta_1, \dots, d\Delta_{p-1} \\ & \times \int_0^\infty \left(\frac{8((1/2)p(p+1)r - r'_u \cdot u)}{(u/(p(p+1)-2) + r)^2} \right. \\ & \left. + 4 \cdot \frac{p(n-p-1) + \text{tr } X^T \nabla \ln f(l)}{u/(p(p+1)-2) + r} \right) \cdot u^{np/2-1} e^{-u/2} du. \end{aligned}$$

Consider the integration

$$I = \int_0^\infty \left(\frac{8((p(p+1)/2)r - r'_u \cdot u)}{(u/(p(p+1)-2) + r)^2} + 4 \cdot \frac{p(n-p-1) + \text{tr } X^T \nabla \ln f(l)}{(u/(p(p+1)-2) + r)} \right) u^{np/2-1} e^{-u/2} du.$$

Let $\psi(u, \Delta_1, \dots, \Delta_{p-1}) = r \cdot u^{-(1/2)p(p+1)}$. Then,

$$\begin{aligned} I = I(\psi) &= -8 \int_0^\infty \frac{\psi'_u u^{(1/2)p(n-p-1)} e^{-u/2} du}{(u^{-(1/2)p(p+1)+1}/(p(p+1)-2) + \psi)^2} \\ &\quad + 4 \int_0^\infty \frac{p(n-p-1) + \text{tr } X^T \nabla \ln f(l)}{(u^{-(1/2)p(p+1)+1}/(p(p+1)-2) + \psi)} \\ &\quad \times e^{-u/2} u^{(p/2)(n-p-1)-1} du. \end{aligned}$$

Using integration by parts, we get

$$\begin{aligned} I(\psi) &= -8 \int_0^\infty \frac{u^{(1/2)pn-p(p+1)} e^{-u/2} du}{2(u^{-(1/2)p(p+1)+1}/(p(p+1)-2) + \psi)^2} \\ &\quad + 4 \int_0^\infty \frac{u + \text{tr } X^T \nabla \ln f(l)}{(u^{-(1/2)p(p+1)+1}/(p(p+1)-2) + \psi)} e^{-u/2} u^{(1/2)p(n-p-1)-1} du. \end{aligned}$$

Considering the problem of maximizing I , we get

$$I(\psi) \leq I(\psi_0) \quad (2.18)$$

where ψ_0 is given by (2.5). Hence

$$\begin{aligned} \Delta(\delta_\beta, 0) &\leq \int \cdots \int_{\mathcal{D}} D(\Delta_1, \dots, \Delta_{p-1}) d\Delta_1, \dots, d\Delta_{p-1} \\ &\quad \times \int_0^\infty \left[\frac{4(u + \text{tr } \nabla \ln f(l)^T X) u^{-p(p+1)/2}}{(u^{-(1/2)p(p+1)+1}/(p(p+1)-2) + \psi_0)} \right. \\ &\quad \left. - \frac{4u^{-p(p+1)+1}}{(u^{-(1/2)p(p+1)+1}/(p(p+1)-2) + \psi_0)^2} \right] u^{+pn/2-1} e^{-u/2} du. \end{aligned}$$

From the above we know that (2.15) holds.

Let

$$\psi_m = \psi_0 \cdot I_{\{\psi_0 \leq m\}} + m \cdot I_{\{\psi_0 > m\}} \quad (2.19)$$

where ψ_0 is given by (2.5). Let

$$\begin{aligned} g_m(X) &= 1 \left/ \left(\frac{\text{tr } S}{p(p+1)-2} + (\text{tr } S)^{(1/2)p(p+1)} \cdot \psi_m \right) \right. \\ &= 1 \left/ \left(\frac{\text{tr } S}{p(p+1)-2} + \beta_m(l) \right) \right. \end{aligned} \quad (2.20)$$

and

$$\delta_{\beta_m}(X) = X + \nabla \ln f(l) - 2Xg_m(X). \quad (2.21)$$

Since $\beta_m(l) = (\text{tr } S)^{p(p+1)/2} \cdot \psi_m$, it is easy to verify that

$$\frac{p(p+1)}{2} \beta_m(l) - \sum_{i=1}^p \frac{\partial \beta_m(l)}{\partial l_i} = -\frac{\partial \psi_m}{\partial u} u^{(1/2)p(p+1)+1} \quad (2.22)$$

where $(u, \Delta_1, \dots, \Delta_{p-1})$ and (l_1, \dots, l_p) are connected by (2.17).

Note that $\beta_m \in B_1$. The technical verification is omitted. According to (2.16) and (2.22),

$$\begin{aligned} \Delta(\delta_{\beta_m}, 0) &= 8E_{\theta=0} \left\{ \frac{-(\partial \psi_m / \partial u) u^{(1/2)p(p+1)+1}}{(u/(p(p+1)-2) + \psi_m u^{(p(p+1))/2})^2} \right\} \\ &\quad + 4E_{\theta=0} \left\{ \frac{p(n-p-1) + \text{tr } \nabla \ln f(l)^T X}{(u/(p(p+1)-2) + \psi_m u^{(p(p+1))/2})} \right\}. \end{aligned}$$

After using integration by parts, we get

$$\begin{aligned} \Delta(\delta_{\beta_m}, 0) &= 4 \int_{\mathcal{D}} \cdots \int D(\Delta_1, \dots, \Delta_{p-1}) d\Delta_1, \dots, d\Delta_{p-1} \\ &\quad \times \int_0^\infty \left(\frac{(u + \text{tr } X^T \nabla \ln f(l)) u^{-(1/2)p(p+1)}}{(u^{-(1/2)p(p+1)+1}/(p(p+1)-2) + \psi_m)} \right. \\ &\quad \left. - \frac{u^{-p(p+1)+1}}{(u^{-(1/2)p(p+1)+1}/(p(p+1)-2) + \psi_m)^2} \right) \\ &\quad \times u^{pn/2-1} e^{-u/2} du. \end{aligned} \quad (2.23)$$

Using the Dominated Convergence Theorem, we get

$$\begin{aligned} \lim_{m \rightarrow +\infty} \Delta(\delta_{\beta_m}, 0) &= 4E_{\theta=0} \left\{ \frac{(\text{tr } S + \text{tr } X^T \nabla \ln f(l)) \text{tr } S^{-(1/2)p(p+1)}}{((\text{tr } S)^{-(1/2)p(p+1)+1}/(p(p+1)-2) + \psi_0)} \right. \\ &\quad \left. - \frac{(\text{tr } S)^{-p(p+1)-1}}{((\text{tr } S)^{-(1/2)p(p+1)+1}/(p(p+1)-2) + \psi_0)^2} \right\}. \end{aligned} \quad (2.24)$$

On the other hand, from the definition of δ^* and δ_{β_m} , it is easy to show that $\lim_{m \rightarrow +\infty} \delta_{\beta_m} = \delta^*$. Hence, after using the Dominated Convergence Theorem, we get

$$\lim_{m \rightarrow +\infty} \Delta(\delta_{\beta_m}, 0) = \Delta(\delta^*, 0). \quad (2.25)$$

Comparing (2.25) with (2.24), we get (2.4). Since we have proved (2.15) it means that (1.6) holds. ■

Using the result of Theorem 1, when $\mu_1(t) = \cdots = \mu_p(t) \equiv 1$, the estimator $\delta^*(X)$ given by (1.9) dominates the crude Efron–Morris estimator $X - (n - p - 1) S^{-1} X$ and is the best at $\theta = 0$ among the class $\{\delta_\beta = X + (n - p - 1) S^{-1} X - 2X/(\text{tr } S/(p(p + 1) - 2) + \beta(l)); \beta \in B_1\}$, when $\mu_1(t) = \cdots = \mu_p(t) \equiv \mu$ given by (1.11), the corresponding estimator $\delta^*(X)$ given by (1.12) dominates the Efron–Morris estimator $(I - (n - p - 1) S^{-1})_+ X$ and is the best at $\theta = 0$ among the class $\{\delta_\beta = (I - (n - p - 1) S^{-1})_+ X - 2X/(\text{tr } S/(p(p + 1) - 2) + \beta); \beta \in B_1\}$.

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